# 12. Functions

Pandas provides a wide range of mathematical functions for performing operations on data frames and series.

Here are some commonly used mathematical functions provided by pandas,

### Aggregating functions,

1. sum(): Computes the sum of values

```
df['A'].sum()
mean(): Computes the mean of values
df['A'].mean()
3. median(): Computes the median of values
df['A'].median()
4. std(): Computes the standard deviation of values
df['A'].std()
5. var(): Computes the variance of values
df['A'].var()
6. min(): Computes the minimum of values
df['A'].min()
```

7. max()\*\*: Computes the maximum of values

```
df['A'].max()
```

8. **count()**: Counts the number of non-null values.

```
df['A'].count()
```

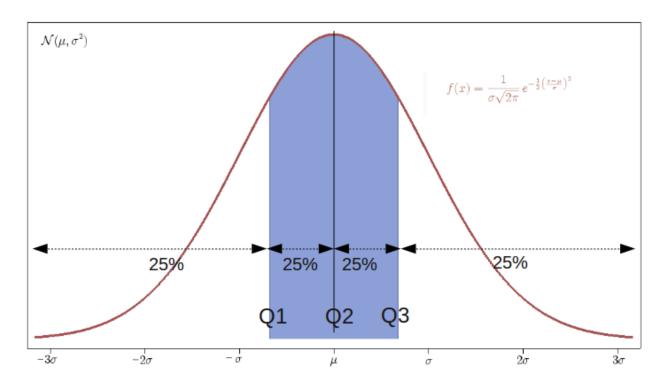
9. **mode()**: Computes the mode(most frequent value)

```
df['A'].mode()
```

10. quantiles(): Computes the quantile of values

```
df['A'].quantile(0.5)
```

In statistics and probability, quantiles are **cut points dividing the range of a probability distribution into continuous intervals with equal probabilities**, or dividing the observations in a sample in the same way.



#### **Cumulative Functions**

1. **cumsum()**: Computes the cumulative sum

```
df['A'].cumsum()
```

A cumulative sum, also known as a **running total**, is the total of a series of data as it increases over time. It's calculated by adding the current value to the sum of all previous

values.

	Α	В	C	D	E
1					
2		Amount	Cumulative		
3		10	10		=SUM(\$B\$3:\$B3)
4		20	30		
5		30	60		
6		40	100		
7		50	150		
8		60	210		

2. cumprod(): Computes the cumulative product

```
df['A'].cumprod()
```

The cumulative product of a sequence is a running products or partial products of a sequence.

The cumulative products of the sequence  $\{a,b,c,...\}$  are a, ab, ab\*c, ....

```
Vector:

3 5 NaN 9 0 NaN

Cummulative product Incude NaN:

3 15 NaN NaN NaN NaN NaN

Cummulative product Exclude NaN:

3 15 15 135 0 0
```

3. **cummin()**: Computes the cumulative minimum

```
df['A'].cummin()
```

The cumulative minimum identifies the minimum value of a column up to and including the current row value.

 In a DataFrame: The cumulative minimum is the lowest value for each column in a DataFrame

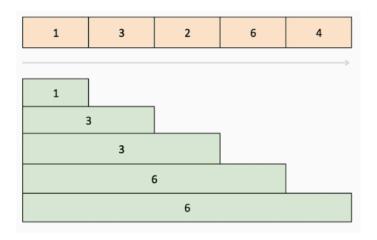
Date   <u>-</u>	Sales ▼	CumulativeMin of Sales 🔻
Jan 2024	100,000	100,000
Feb 2024	95,000	95,000
Mar 2024	110,000	95,000
Apr 2024	80,000	80,000
May 2024	90,000	80,000
Jun 2024	75,000	75,000
Jul 2024	100,000	75,000
Aug 2024	105,000	75,000
Sep 2024	95,000	75,000
Oct 2024	120,000	75,000
Nov 2024	125,000	75,000
Dec 2024	120,000	75,000

4. **cummax()**: Computes the cumulative maximum

```
df['A'].cummax()
```

Cumulative maximum refers to the maximum value of a column, including the current row value.

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**Cumulative Max** 

# **Descriptive Statistics**

1. **describe()**: Generates descriptive statistics

```
df.describe()
```

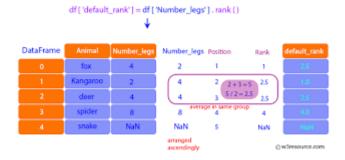
If the DataFrame contains numerical data, the description contains these information for each column: count - The number of not-empty values. mean - The average (mean) value. std - The standard deviation.

```
import pandas as pd
city_names = pd.Series(['San Francisco', 'San Jose', 'Sacramento'])
cities.describe()
        Population Area square miles Population density
count 3.000000e+00
                          3.000000
                                           3.000000
mean 7.844843e+05
                         107.106667
                                         9632.392763
      2.717477e+05
                                         7420.091623
                         65.316346
      4.851990e+05
                         46.870000
                                         4955.055147
 min
      6.688340e+05
                                         5354.616454
 25%
                         72.395000
 50%
      8.524690e+05
                         97.920000
                                         5754.177760
      9.341270e+05
                         137.225000
                                        11971.061570
     1.015785e+06
                         176.530000
                                        18187.945381
 max
```

# **Ranking Functions**

1. rank(): Computes numerical data ranks

```
df['A'].rank()
```



#### **Difference Functions**

1. **diff()**: Computes the difference of a DataFrame element compared with another element in the DataFrame(Default is element in the previous row)

```
df['A'].diff()
```

```
main.py
# Example: diff
import pandas as pd
# Create a dataframe
df = pd.DataFrame({'A': [1, 2, 3, 4, 5, 6],
                     'B': [1, 1, 2, 3, 5, 8],
                        'C': [1, 4, 9, 16, 25, 36]})
# Syntax: DataFrame.diff(periods=1, axis=0)
# Calculate the difference between the current and a shifted row
print(df.diff())
           NaN
                  NaN
      NaN
                  5.0
                 11.0
```

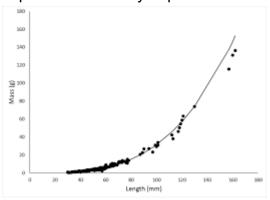
## **Expanding Transformation**

1. **expanding()**: Provides expanding transformations.

```
df['A'].expanding().sum()
```

In statistics, an "expanding transformation" generally refers to a data transformation technique where the original data points are replaced with a new set of values that are progressively "stretched out" or expanded across a wider range, often achieved by applying a mathematical function that accentuates differences between data points, typically used to

improve the visibility of patterns or relationships in highly concentrated data sets.

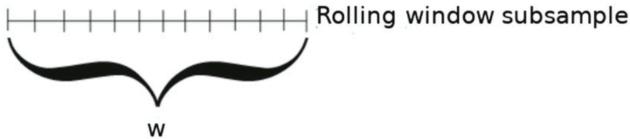


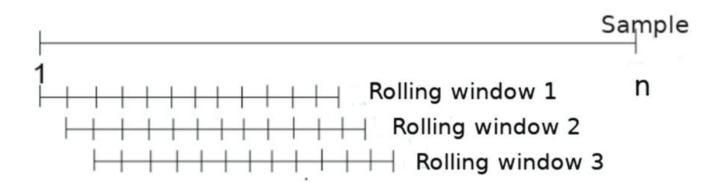
## **Rolling window functions**

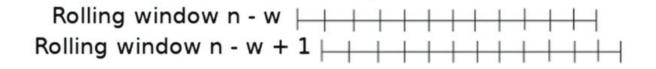
1. rolling(): Provides rolling window calculations.

```
df['A'].rolling(window=3).mean()
```

Rolling window calculations are a way to calculate statistics that move with data, such as averages, sums, or maximums.







# Other mathematical functions

1. abs(): Computes the absolute values elementwise

```
df['A'].abs()
```

2. round(): Rounds values to a specified number of decimal places

```
df['A'].round(2)
```

3. **pow()**: Computes the power of elements

```
df['A'].pow(2)
```

3. **pct\_change()**: Computes the percentage change between the current and a prior element

```
df['A'].pct_change()
```

3. clip(): Trims values at input threshold(s)

```
df['A'].clip(lower=0, upper=10)
```